

# The Battery-Discharge-Model: A Class of Stochastic Finite Automata to Simulate Multidimensional Continued Fraction Expansion

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**ABSTRACT** We define an infinite stochastic state machine, the *Battery-Discharge-Model* (BDM), which simulates the behaviour of linear and jump complexity of the continued fraction expansion of multidimensional formal power series, a relevant security measure in the cryptanalysis of stream ciphers.

We also obtain finite approximations to the infinite BDM, where *polynomially* many states suffice to approximate with an *exponentially small* error the probabilities and averages for linear and jump complexity of  $M$ -multisequences of length  $n$  over the finite field  $\mathbb{F}_q$ , for *any*  $M$ ,  $n$ ,  $q$ .

## Introduction

In cryptography two important measures of sequence complexity are the linear and jump complexity, dealing with the continued fraction expansion of the sequence seen as formal power series over some finite field  $\mathbb{F}_q$ . While both complexities are well understood for single sequences ([3],[6]), a current topic is to generalize these notions to multisequences ( $M$  streams of symbols in parallel) with first results for  $M = 2, q = 2$  given in [2].

In Section I we suggest an infinite recurrent stochastic automaton and finite approximations, the *Battery-Discharge-Model* that simulates the continued fraction expansion (proof in Section II) and thus (Section III) answers questions about linear and jump complexity for every  $M$  (“multi”-ness),  $q$  (order of finite field), and  $n$  (length of sequence).

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## I. The Battery–Discharge–Model

In this first part, we develop in three steps an infinite stochastic automaton, the *Battery–Discharge–Model* and a family of finite approximations.

### a) Model without discharge

We fix a number  $M \in \mathbb{N}$  and then have  $M$  batteries, each holding a charge  $b_i \in \mathbb{Z}, 1 \leq i \leq M$ , and a drain  $d \in \mathbb{Z}$ , hence an infinite number of possible states. The initial state is  $d = b_i = 0$ .

The model cyclically runs through  $M + 1$  main cycles  $T = 0, \dots, M$ . At each transition  $T \rightarrow T + 1$  for  $T = 0, \dots, M - 1$ , the drain is decremented,  $d := d - 1$ , whereas the batteries do not change. At the transition from  $T = M$  to  $T = 0$ , all batteries are incremented,  $b_i := b_i + 1, 1 \leq i \leq M$ , whereas  $d$  remains constant. With the initial condition  $d = b_i = 0$ , we thus have

$$T + d + \sum_{i=1}^M b_i \equiv 0 \quad (\text{invariant})$$

Writing the state in the form  $(b_1, \dots, b_M; d)_T$ , we obtain the following behaviour for this model:  $(0, \dots, 0; 0)_0 \rightarrow (0, \dots, 0; -1)_1 \rightarrow (0, \dots, 0; -2)_2 \rightarrow \dots \rightarrow (0, \dots, 0; -M)_M \rightarrow (1, \dots, 1; -M)_0 \rightarrow (1, \dots, 1; -M - 1)_1 \rightarrow (1, \dots, 1; -M - 2)_2 \rightarrow \dots \rightarrow (1, \dots, 1; -2M)_M \rightarrow (2, \dots, 2; -2M)_0 \rightarrow \dots$

### b) Model with discharge

Each of the  $M + 1$  major cycles now is divided into  $M + 1$  subcycles  $t = *, 1, \dots, M$ . Subcycle  $*$  shows the result of decrementing  $d$  or incrementing the  $b_i$ , whereas during subcycle  $t, t = 1, \dots, M$ , battery  $b_t$  may *discharge* into the drain, provided it has high enough potential that is  $b_t > d$ . In this case the excess charge goes from  $b_t$  to the drain, amounting to an interchange  $d \leftrightarrow b_t$  of values, thus maintaining the invariant.

The behaviour with discharge is as follows (for illustration we use  $M = 3$ ): The underlined battery is the one,  $b_t$ , corresponding to the subcycle. We show the result at the *end* of the subcycle. In case of a discharge, (the new)  $b_t$  is in boldface:

$$\begin{aligned} (b_1, b_2, b_3; d)_{T,t} &= (0, 0, 0; 0)_{0,*} \rightarrow (\underline{0}, 0, 0; 0)_{0,1} \rightarrow (0, \underline{0}, 0; 0)_{0,2} \rightarrow (0, 0, \underline{0}; 0)_{0,3} \rightarrow \\ &(0, 0, 0; -1)_{1,*} \rightarrow (\underline{-1}, 0, 0; 0)_{1,1} \rightarrow (-1, \underline{0}, 0; 0)_{1,2} \rightarrow (-1, 0, \underline{0}; 0)_{1,3} \rightarrow \\ &(-1, 0, 0; -1)_{2,*} \rightarrow (\underline{-1}, 0, 0; -1)_{2,1} \rightarrow (-1, \underline{-1}, 0; 0)_{2,2} \rightarrow (-1, -1, \underline{0}; 0)_{2,3} \rightarrow \\ &(-1, -1, 0; -1)_{3,*} \rightarrow (\underline{-1}, -1, 0; -1)_{3,1} \rightarrow (-1, \underline{-1}, 0; -1)_{3,2} \rightarrow (-1, -1, \underline{-1}; 0)_{3,3} \rightarrow \end{aligned}$$

$$(0, 0, 0; 0)_{0,*} \rightarrow (\underline{0}, 0, 0; 0)_{0,1} \rightarrow (0, \underline{0}, 0; 0)_{0,2} \rightarrow (0, 0, \underline{0}; 0)_{0,3} \rightarrow \dots$$

The behaviour of this model I.b) is (purely) periodic.

### c) Model with discharge and inhibition

Finally we introduce the stochastic element of temporary inhibition of a battery (observe that in I.a) the batteries never discharge, in I.b) always):

Let  $\frac{1}{q}$ , with  $0 < \frac{1}{q} \leq 1$ , be the probability that a battery with  $b_t > d$  will be *inhibited* and thus will not discharge. We now have a stochastic behaviour:  $(0, 0, 0; 0)_{0,*} \rightarrow \dots (0, 0, 0; -1)_{1,*}$  with probability 1, but  $(0, 0, 0; -1)_{1,*} \rightarrow (-1, 0, 0; 0)_{1,1}$  with probability  $\frac{q-1}{q}$ , and  $(0, 0, 0; -1)_{1,*} \rightarrow (\underline{0}, 0, 0; -1)_{1,1}$  with probability  $\frac{1}{q}$ , and then battery 2 may be inhibited or not etc.

A more involved example: From  $(0, 2, 1; -4)_{1,*}$  within cycle  $T = 1$ , six outcomes are possible (“D” = discharge, “I” = inhibition, “−” =  $b_i \leq d$ ):

$(b_1, b_2, b_3; d)_{1,*}$	$b_1$	$b_2$	$b_3$	Prob	$(d; b_1, b_2, b_3)_{1,3}$	$\rightarrow$	$(d; b_1, b_2, b_3)_{2,*}$
$(0, 2, 1; -4)$	D	D	−	$(q-1)^2/q^2$	$(-4, 0, 1; 2)$	$\rightarrow$	$(-4, 0, 1; 1)$
$(0, 2, 1; -4)$	D	I	D	$(q-1)^2/q^3$	$(-4, 2, 0; 1)$	$\rightarrow$	$(-4, 2, 0; 0)$
$(0, 2, 1; -4)$	D	I	I	$(q-1)/q^3$	$(-4, 2, 1; 0)$	$\rightarrow$	$(-4, 2, 1; -1)$
$(0, 2, 1; -4)$	I	D	−	$(q-1)/q^2$	$(0, -4, 1; 2)$	$\rightarrow$	$(0, -4, 1; 1)$
$(0, 2, 1; -4)$	I	I	D	$(q-1)/q^3$	$(0, 2, -4; 1)$	$\rightarrow$	$(0, 2, -4; 0)$
$(0, 2, 1; -4)$	I	I	I	$1/q^3$	$(0, 2, 1; -4)$	$\rightarrow$	$(0, 2, 1; -5)$

For instance, the transition of the second line consists of these steps:

$$(0, 2, 1; -4) \xrightarrow{D; \frac{q-1}{q}} (\underline{-4}, 2, 1; 0) \xrightarrow{I; \frac{1}{q}} (-4, \underline{2}, 1; 0) \xrightarrow{D; \frac{q-1}{q}} (-4, 2, \underline{0}; 1) \xrightarrow{d:=-d-1} (-4, 2, 0; 0)$$

### d) Properties of the full model<sup>2</sup>

This is our full model. We shall use only the states at timesteps  $(T, *)$  and let the transition probability take care of the events (discharge, inhibition) during subcycles  $t = 1, \dots, M$ . The state set is isomorphic to  $\mathbb{Z}^M \times \{0, \dots, M\}$ , where the  $M$  battery levels and the main cycle are given and the drain value results implicitly from  $d = -T - \sum_{i=1}^M b_i$ .

Every transition probability is of the form  $\frac{(q-1)^a}{q^b}$  for  $b$  batteries with  $b_i > d$  at their subcycle, with  $a$  discharges and  $b - a$  inhibitions. So, all transition (and state) probabilities are polynomial functions in  $q$  by construction. The values

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<sup>2</sup>..which of course could be called the Butterfly-Dinner-Model, where  $M$  flowers  $f_i$  are visited in turn by a butterfly, dining from the supplied nectar, whenever the level exceeds that of the butterfly, with probability  $\frac{1}{q}$  that the flower has its petals closed...

for  $b_i, d$  in the successor state  $(T + 1, *)$  already incorporate the decrement of  $d$  or increment of the  $b_i$ .

Let  $Q_T$  be the (infinite) set of all states in main cycle  $T, T = 0, \dots, M$ . We adjoin a class  $K \in \mathbb{N}_0$  to each state as follows: State  $(0, \dots, 0; 0)_{0,*}$  and all states reachable without inhibition from here are in class 0 (these are just the states  $(\dots)_{T,*}$  of model I.b). All states reachable from a class  $K$  with  $i \in \mathbb{N}_0$  inhibitions belong to class  $K + i$ . If a state can be reached in different ways (number of inhibitions), the *smallest* such class number applies. The rationale for these classes is:

i) Numerical evidence shows that in the stationary distribution of the infinite model in each cycle  $T$ , *every* state of class  $K$  occurs with probability *exactly*  $q^{-K}$  times the probability of the (unique) state in class 0.

ii) For  $q \rightarrow \infty$ , almost always we are in the states of class 0. Hence, restricting the (infinite) state set  $\mathbb{Z}^M \times \{0, \dots, M\}$  to those states within classes  $0 \dots K_0$ , for some fixed accuracy  $K_0 \in \mathbb{N}_0$ , we obtain – at least for large  $q$  – a fairly good approximation to the infinite model. In this case, for a state in class  $K$ , the model allows at most  $K_0 - K$  more inhibitions and thus the  $(K_0 - K + 1)$ -st battery with  $b > d$  has to discharge (with probability 1).

iii) Also, simulation results for  $K_0$  up to 120 show that the number of states in class  $K$  for the unbounded model and for each main cycle  $T$  is  $p_K(M)$ , the number of partitions of  $K$  into at most  $M$  parts or – what is the same – into parts of size at most  $M$ .  $p_K(M)$  grows as  $\approx \frac{K^{M-1}}{M!(M-1)!}$  asymptotically for fixed  $M$  and  $K \rightarrow \infty$  (see INRIA [8] and Sloane's [9] integer sequences). Thus the bounded finite models have an overall number of states  $(M+1) \cdot \sum_{K=0}^{K_0} p_K(M)$ . Furthermore, let us define  $\mathcal{P}(M, q) = \sum_{K \in \mathbb{N}_0} p_K(M) \cdot q^{-K} = \sum_{s \in Q_T} q^{-K(s)}$  (the same for every  $T$ ). Then a state of class  $K$  has probability  $q^{-K} / \mathcal{P}(M, q)$ .

**Example** The bounded model for  $M = 3$  and  $K_0 = 2$  consists of  $(3 + 1) \cdot \sum_{K=0}^2 p_K(3) = 16$  states (one state each in class 0 and 1, two states in class 2, for each  $T$ ). Since  $\mathcal{P}(3, q) = q^3 / ((q - 1)^2(q + 1))$ , already this *very* limited model accounts for a share  $(1 \cdot q^0 + 1 \cdot q^1 + 2 \cdot q^2) / \mathcal{P}(M, q)$  of the stationary probability distribution of the unbounded model, which is 75% for  $q = 2$ , 99.6% for  $q = 8$  etc.

Here we give all states with class  $K \leq K_0 = 2$ , belonging to subcycle  $(T, *)$  together with all of their successor states in subcycle  $(T + 1, *)$  and the respective transition probability. In the first line *e.g.*, we have a probability of  $1/q^2$  to go to  $(0, 0, -1; 0)$  instead of  $(q - 1)/q^3$  (for 2 inhibitions and 1 discharge), since battery 2 *has* to discharge to keep within  $K \leq K_0$ , state  $(0, 0, 0; -1)$  does not appear in this bounded model.

$T$	$K$	$(b_1, b_2, b_3; d)$	Nextstates : Probability
0	0	$(0, 0, 0; 0)$	$(-1, 0, 0; 0) : \frac{q-1}{q}, (0, -1, 0; 0) : \frac{q-1}{q^2}, (0, 0, -1; 0) : \frac{1}{q^2}$
0	1	$(-1, 0, 0; 1)$	$(-1, 0, 0; 0) : 1$
0	2	$(0, -1, 0; 1)$	$(0, -1, 0; 0) : 1$
0	2	$(-1, 0, 1; 0)$	$(-1, -1, 0; 1) : 1$
1	0	$(-1, 0, 0; 0)$	$(-1, -1, 0; 0) : \frac{q-1}{q}, (-1, 0, -1; 0) : \frac{q-1}{q^2}, (-1, 0, 0; -1) : \frac{1}{q^2}$
1	1	$(0, -1, 0; 0)$	$(-1, -1, 0; 0) : \frac{q-1}{q}, (0, -1, -1; 0) : \frac{1}{q}$
1	2	$(0, 0, -1; 0)$	$(-1, 0, -1; 0) : \frac{q-1}{q}, (0, -1, -1; 0) : \frac{1}{q}$
1	2	$(-1, -1, 0; 1)$	$(-1, -1, 0; 0) : 1$
2	0	$(-1, -1, 0; 0)$	$(-1, -1, -1; 0) : \frac{q-1}{q}, (-1, -1, 0; -1) : \frac{1}{q}$
2	1	$(-1, 0, -1; 0)$	$(-1, -1, -1; 0) : \frac{q-1}{q}, (-1, 0, -1; -1) : \frac{1}{q}$
2	2	$(-1, 0, 0; -1)$	$(-2, -1, 0; 0) : 1$
2	2	$(0, -1, -1; 0)$	$(-1, -1, -1; 0) : 1$
3	0	$(-1, -1, -1; 0)$	$(0, 0, 0; 0) : 1$
3	1	$(-1, -1, 0; -1)$	$(-1, 0, 0; 1) : \frac{(q-1)^2}{q^2}, (-1, 0, 1; 0) : \frac{q-1}{q^2}, (0, -1, 0; 1) : \frac{1}{q}$
3	2	$(-1, 0, -1; -1)$	$(-1, 0, 0; 1) : \frac{q-1}{q}, (0, -1, 0; 1) : \frac{1}{q}$
3	2	$(-2, -1, 0; 0)$	$(-1, 0, 0; 1) : \frac{q-1}{q}, (-1, 0, 1; 0) : \frac{1}{q}$

## II. The BDM and Continued Fraction Expansion

We now apply the BDM to obtain precise values about the behaviour of the linear and jump complexity of multisequences: Let  $G_t(a) = \sum_{i=1}^{\infty} a_{t,i} x^{-i} \in \mathbb{F}_q[[x^{-1}]]$ ,  $t = 1, \dots, M$  be  $M$  formal power series over the finite field  $\mathbb{F}_q$ .

The linear complexity of  $(G_t(a) \mid 1 \leq t \leq M)$  at  $n$  is defined as the smallest degree of a polynomial  $v(x)$ , such that there are some polynomials  $u_t(x)$ ,  $1 \leq t \leq M$  with  $G_t(x) = \frac{u_t(x)}{v(x)} + O(x^{-(n+1)})$ . The jump complexity in turn counts, how often this smallest degree has changed (increased) until step  $n$  (see [4][5]).

We derive these complexities from our BDM, using its equivalence to the multi-Strict Continued Fraction Algorithm (m-SCFA) of Dai and Feng [2].

The m-SCFA uses the following variables to describe the state:

$n$ , the timestep

$d =: d_{\text{SCFA}}$ , the degree of  $v$ , the current approximation denominator

$w_t$ ,  $1 \leq t \leq M$ , a “degree deviation” of  $u_t(x)$  at sequence  $t$

Our BDM uses the equivalent variables:

(\*)  $T$ , timestep, with  $T \equiv n \bmod (M+1)$

(\*\*)  $d =: d_{\text{BDM}}$ ,  $d_{\text{BDM}} = d_{\text{SCFA}} - \lceil \frac{n \cdot M}{M+1} \rceil$ , the deviation of  $\deg(v)$  from its typical value,

$(***)$   $b_t, b_t = \lfloor \frac{n}{M+1} \rfloor - w_t, 1 \leq t \leq M$ , the battery levels.

Observe that initially (at  $n = T = 0$ )  $d_{\text{SCFA}} = d_{\text{BDM}} = w_t = b_t = 0, \forall t$ , so both models coincide according to equivalences  $(*)$  to  $(***)$ .

Let us first consider the timestep  $n$ , main cycle  $T$ , at subcycle  $*$ :

Assuming  $d_{\text{SCFA}}, w_t$  fix with  $n \rightarrow n + 1$  we must have the new values

$$d_{\text{BDM}}^+ \stackrel{(**)!}{=} d_{\text{SCFA}} - \left\lfloor \frac{(n+1) \cdot M}{M+1} \right\rfloor = d_{\text{SCFA}} - \left\lfloor \frac{n \cdot M}{M+1} \right\rfloor - \varepsilon = d_{\text{BDM}} - \varepsilon$$

where  $\varepsilon = 0$  for  $n+1 \equiv 0 \pmod{M+1}$  and 1 otherwise, and

$$b_t^+ \stackrel{(***)!}{=} \left\lfloor \frac{n+1}{M+1} \right\rfloor - w_t = \left\lfloor \frac{n}{M+1} \right\rfloor + \varepsilon - w_t = b_t + \varepsilon, \forall t$$

where  $\varepsilon = 1$  for  $n+1 \equiv 0 \pmod{M+1}$  and 0 otherwise. This corresponds to incrementing the  $b_t$ 's for  $T \equiv M \rightarrow 0$  and otherwise decrementing  $d$ .

Now, within the  $M$  subcycles  $t = 1, \dots, M$  we consider four cases, according to a “discrepancy”  $\delta$  of the m-SCFA (the deviation between the formal power series and the approximation by  $u_t(x)/v(x)$ ) and the values of  $n, d, w_t$ :

	m-SCFA	[2, Thm. 2]	BDM Case
1	$\delta = 0$ and $n - d_{\text{SCFA}} \leq w_t$	a	level too low, “ $-$ ”
2	$\delta \neq 0$ and $n - d_{\text{SCFA}} \leq w_t$	c	level too low, “ $-$ ”
3	$\delta = 0$ and $n - d_{\text{SCFA}} > w_t$	a	inhibition “I”
4	$\delta \neq 0$ and $n - d_{\text{SCFA}} > w_t$	b	discharge “D”

First note that  $n - d_{\text{SCFA}} > w_t \Leftrightarrow n - (\lfloor \frac{n \cdot M}{M+1} \rfloor + d_{\text{BDM}}) > \lfloor \frac{n}{M+1} \rfloor - b_t \Leftrightarrow b_t > d_{\text{BDM}}$  corresponds to cases 3 and 4, that is discharge or inhibition. We model a discrepancy value  $\delta = 0$  by the probability of inhibition  $1/q$ , according to the following proposition about the even distribution of discrepancy values.

**Proposition** *In any given position  $(m, n), 1 \leq m \leq M, n \in \mathbb{N}$  of the formal power series, exactly one choice for the next symbol  $a_{m,n}$  will yield a discrepancy  $\delta = 0$ , all other  $q - 1$  symbols from  $\mathbb{F}_q$  result in some  $\delta \neq 0$ .*

**Proof:** The current approximation  $u_m^{(m,n)}(x)/v^{(m,n)}(x)$  determines exactly one approximating coefficient sequence for the  $m$ -th formal power series  $G_m$ . The (only) corresponding symbol belongs to  $\delta = 0$ .  $\square$

In fact, for every position  $(m, n)$ , each discrepancy value  $\delta \in \mathbb{F}_q$  occurs exactly once for some  $a_{m,n} \in \mathbb{F}_q$ , in other words (compare [1][5] for  $M = 1$ ):

**Fact** *The Generalized Berlekamp–Massey–Algorithm (GBMA) and the multi-Strict Continued Fraction Algorithm (sCFA) induce an isometry on  $(\mathbb{F}_q^M)^\omega$ .*

Concerning the update of the  $d_{\text{SCFA}}$  and  $w_t$  values, in cases 1 to 3 nothing happens, neither in the m-SCFA, nor in the BDM. In case 4 the updated

values in [2] are  $d_{\text{SCFA}}^+ = n - w_t$  and  $w_t^+ = n - d_{\text{SCFA}}$ , thus our BDM must set:

$$d_{\text{BDM}}^+ \stackrel{(**)}{=} d_{\text{SCFA}}^+ - \left\lfloor \frac{n \cdot M}{M+1} \right\rfloor \stackrel{[2]}{=} (n - w_t) - \left\lfloor \frac{n \cdot M}{M+1} \right\rfloor \stackrel{***}{=} \left\lfloor \frac{n}{M+1} \right\rfloor + b_t - \left\lfloor \frac{n}{M+1} \right\rfloor = b_t$$

and

$$b_t^+ \stackrel{(***)}{=} \left\lfloor \frac{n}{M+1} \right\rfloor - w_t^+ \stackrel{[2]}{=} \left\lfloor \frac{n}{M+1} \right\rfloor - (n - d_{\text{SCFA}}) \stackrel{(**)}{=} - \left\lfloor \frac{n \cdot M}{M+1} \right\rfloor + d_{\text{BDM}} + \left\lfloor \frac{n \cdot M}{M+1} \right\rfloor = d_{\text{BDM}},$$

that is interchange of  $d_{\text{BDM}}$  with  $b_t$ , as takes place in a discharge.

Finally, our transition probability over all  $M$  subcycles of  $D, I$  or  $-$  is the product  $\left(\frac{q-1}{q}\right)^{\#D} \cdot \left(\frac{1}{q}\right)^{\#I} \cdot \left(\frac{q}{q}\right)^{\#-}$  (where  $\#D + \#I + \#- = M$ ), corresponding to  $(q-1)^{\#D} 1^{\#I} q^{\#-}$  different  $M$ -tuples of symbols in row  $n$  of the  $M$  formal power series ( $\#D$  times  $\delta \neq 0$ ,  $\#I$  times  $\delta = 0$ ,  $\#-$  times *any* symbol from  $\mathbb{F}_q$ ).

### III. Numerical Results about Multidimensional Linear and Jump Complexity

We have an infinite model and finite approximations that simulate the behaviour of the multidimensional continued fraction expansion algorithm: The drain  $d$  corresponds to the linear complexity deviation  $d = \deg(v) - \left\lceil \frac{n \cdot M}{M+1} \right\rceil$ , whereas each “D” in a transition corresponds to a jump by a height  $b_t - d$ .

We start at time 0 with a probability distribution of  $pr(0, \dots, 0; 0) = 1$ , zero everywhere else, and run the state transition matrix until reaching the stationary equilibrium.

#### a) Linear complexity deviation

The average linear complexity deviation in level  $T$  is  $(\mathcal{P}(M, q)$  as in I.d):

$$\bar{d}(M, T) = \frac{\sum_{s \in Q_T} q^{-K(s)} \cdot d(s)}{\sum_{s \in Q_T} q^{-K(s)}} = \frac{\sum_{s \in Q_T} q^{-K(s)} \cdot d(s)}{\mathcal{P}(M, q)}, \quad T = 0, \dots, M.$$

Also, we have  $\bar{d}(M) = \sum_{T=0}^M \bar{d}(M, T) / (M+1)$  as average over all  $T$ .

$\bar{d}(M)$  turns out to be zero for all  $M$  and  $q$ , another argument for our choice of  $\deg(v) \approx \left\lceil \frac{n \cdot M}{M+1} \right\rceil$  as “typical” behaviour.

The probability that the degree deviation has a certain value  $d_0$ , for some  $T$ , is  $pr(d = d_0)_{(M, T)} = \left( \sum_{s \in Q_T, d(s)=d_0} q^{-K(s)} \right) / \mathcal{P}(M, q)$  and we set  $pr(d = d_0)_M = \frac{1}{M+1} \sum_{T=0}^M pr(d = d_0)_{(M, T)}$ . We have  $pr(d = d_0)_M = pr(d = -d_0)_M$ .

The general (in  $q$ ) formula for  $\bar{d}(M, T)$  is  $\bar{d}(1, 0) = -\bar{d}(1, 1) = q/(q+1)^2$ ,  $\bar{d}(2, 0) = -\bar{d}(2, 2) = (q^5 + q^4 - q^3 + q^2 + q)/(q^3 + 1)(q^2 + q + 1)^2$ ,  $\bar{d}(2, 1) = 0$ , and in general  $\bar{d}(M, T) = -\bar{d}(M, M - T)$  for  $0 \leq T \leq M$ , leading to  $\bar{d}(M) = 0$  (all this by numerical evidence). For  $q = 2$ , we obtain

$M$	$\bar{d}(M, 0)$	$\bar{d}(M, 1)$	$\bar{d}(M, 2)$	$\bar{d}(M, 3)$	$\bar{d}(M, 4)$
1	0.222222	-0.222222			
2	0.312925	0	-0.312925		
3	0.352021	0.10806	-0.10806	-0.352021	
4	0.370890	0.163309	0	-0.163309	-0.370890
5	0.380275	0.191638	0.0572067	-0.0572067	-0.191638
6	0.384972	0.206045	0.0868915	0	-0.0868915
7	0.387277	0.213270	0.1019991	0.0297877	-0.0297877
8	0.388441	0.216919	0.1096532	0.0450379	0

$M$	$p(d = 0)$	$p(d = \pm 1)$	$p(d = \pm 2)$	$p(d = \pm 3)$	$p(d = \pm 4)$	$p(d = \pm 5)$
1	0.5	0.1875	0.046875	0.011719	0.002930	0.000732
2	0.55	0.19414	0.026959	0.03412	0.000427	$5.34e - 5$
3	0.61920	0.176843	0.012701	0.0008006	$5.0066e - 5$	$3.1292e - 6$

For  $q = 100$  (remember that our model requires only  $2 \leq q \in \mathbb{R}$ ), the values for  $d(M, T)$  (and similar for all the other results) suggest formal power series in  $q^{-1}$ , as such valid for any  $q$  (the dots separate the powers of  $q^{-1}$ ):

$M$	$\bar{d}(M, 0)$	$\bar{d}(M, 1)$
1	0,00.98.02.96.04.94.06.92	
1	0,00.99.97.03.01.94.04.02.86	0
3	0,00.99.98.98.03.01.98.92.99	0,00.00.98.99.98.03.00.00.96.9

For  $M = 1$  and 2 the closed form was already given, for  $M = 3$  we obtain:

$$\bar{d}(3, 0) = 1q^{-1} + 0q^{-2} - 1q^{-3} - 2q^{-4} + 3q^{-5} + 2q^{-6} - 1q^{-7} - 7q^{-8} \pm \dots$$

$$\bar{d}(3, 1) = 0q^{-1} + 1q^{-2} - 1q^{-3} + 0q^{-4} - 2q^{-5} + 3q^{-6} + 0q^{-7} + 1q^{-8} - 3q^{-9} \pm \dots$$

## b) Jump complexity

The jump complexity counts how many discharges occur, and with which height  $b_t - d$ . Let  $s_1 \xrightarrow{t} s_2$  with  $t \in \{I, D, -\}^M$  be some transition, where  $t$  denotes the actions at the  $M$  batteries. Let  $t_I, t_D, t_-$  be the respective number of symbols  $I, D$ , and  $-$  in  $t$ , then  $t$  has overall probability  $\frac{q^{-K(s_1)}}{\mathcal{P}(M, q)} \cdot \frac{(q-1)^{t_D}}{q^{t_I+t_D}}$ . Hence, we have an average jump complexity per time unit of

$$\bar{J}(T) = \sum_{s_1 \xrightarrow{t} s_2, s_1 \in Q_T, s_2 \in Q_{T+1}} t_D \cdot \frac{q^{-K(s_1)}}{\mathcal{P}(M, q)} \cdot (q-1)^{t_D} q^{-(t_I+t_D)},$$



hence up to  $n$  an expected average of  $n \cdot \bar{J} := n \cdot \frac{1}{M+1} \sum_{T=0}^M \bar{J}(T)$  jumps.

Also, we calculate how many jumps by height  $h \in \mathbb{N}$  occur on average as:

$$\overline{JH}(h) = \frac{1}{M+1} \sum_{T=0}^M \sum_{\substack{s_1 \xrightarrow{t} s_2 \\ s_1 \in Q_T, s_2 \in Q_{T+1}}} |\{i \mid b_i - d = h \text{ in } t\}| \cdot \frac{q^{-K(s_1)}}{\mathcal{P}(M, q)} \cdot (q-1)^{t_D} q^{-t_I - t_D}.$$

Again, we list some values and also have a closed formula for  $M = 1$  and 2.

$q$	$M$	$\bar{J}$	$\overline{JH}(1)$	$\overline{JH}(2)$	$\overline{JH}(3)$	$\overline{JH}(4)$	$\overline{JH}(5)$
2	1	0.25	0.125	0.0625	0.03125	0.015625	0.007813
	2	0.44444	0.29167	0.10417	0.03385	0.10417	0.003092
	3	0.58929	0.45786	0.10699	0.02023	0.003456	0.000549
	4	0.69333	0.59742	0.08610	0.00895	0.000794	0.000065
	5	0.76613	0.70254	0.06014	0.00329	0.000149	0.000006
	6	0.81633	0.77665	0.03856	0.00109	0.000025	0.000001
$q$	$M$	$\bar{J}$	$\overline{JH}(1)$	$\overline{JH}(2)$	$\overline{JH}(3)$		
100	1	0.495	0.49005	0.0049005	0.000049005		
	2	0.6665346534	0.66640266	0.00013197373	...		

That is for  $M = 1$  we have  $\bar{J} = \frac{1}{2} - \frac{1}{q} + \frac{1}{2q}$  and  $\overline{JH}(h) = q^{-h+1} \cdot \left( \frac{1}{2} - \frac{1}{q} + \frac{1}{2q^2} \right)$ , and for  $M = 2$  we obtain by evaluating for several  $q$ :  $\bar{J} = \frac{2}{3} - \frac{4}{3q(q+1)}$ . Observe that for  $q \rightarrow \infty$  and any  $M$ , we have  $\bar{J} = \frac{M}{M+1}$ , according to model I.b).

## Open Problems:

1. Show *algebraically* that  $\forall s_1, s_2 \in Q_T$  we have  $\frac{pr(s_1)}{pr(s_2)} = q^{-K(s_1)+K(s_2)}$ , where  $K(s)$  is defined via the number of inhibitions from  $(0, \dots, 0; 0)$ .
2. Show *algebraically* that  $|\{s \in Q_T \mid K(s) = K\}| = p_K(M)$  for all  $K, T, M$ .
3. Give a closed form for the coefficients of all the new formal power series in  $\mathbb{Z}[[q^{-1}]]$  occuring in this paper.

## Conclusion

We developed a model of multidimensional linear and jump complexity, using a stochastic infinite state machine, which is selfsimilar on the time axis, folding back time mod  $M + 1$  onto itself.

Fixing an arbitrary good accuracy level  $K_0$ , we obtain a *finite* model that approximates with an *exponentially* small (in  $K_0$ ) error, using only *polynomially*

many states.

We derived values for linear and jump complexity of multisequences in the average case and probabilities for deviations from that case.

The whole theory is valid for *any*  $q$  (order of finite field), *any*  $M$  (number of sequences) and *any* timestep  $n$ , We have numerical results for  $M$  up to 8,  $n \rightarrow \infty$ , and any  $q$ , extending considerably the range of known results.

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**Appendix** State counts for  $M = 3$ ,  $T = 0$ ,  $K = 0, 1, \dots, 50$  (vertical) and  $d = -10, -9, \dots, 9$  (horizontal),  $d = 0$  within dots. The last column is  $p_K(M)$ .

[illegible]